```
In [1]: import numpy as np
  import pandas as pd
  import matplotlib.pyplot as plt
  from scipy import interpolate
  from mpl_toolkits.mplot3d import Axes3D
  from collections import Counter
  import numpy as np
  from scipy.spatial import cKDTree
```

Question 1

```
In [5]: # Differential Equation
def diffunc(x,y):
    return 1/(x**2 * (1-y))
```

Question 1.b

```
In [6]: # Euler Method
def euler(x0, y, h, x):
    # Iterating till x
    while x0 > x:
        y = y + h * diffunc(x0, y)
        x0 = x0 + h
    print("Approximate solution at x = ", x, " is ", "%.6f"% y)
euler(1, -1, -0.05, 0)
euler(1, -1, -0.05, 0.05)
```

Approximate solution at x = 0 is -8.124934Approximate solution at x = 0.05 is -4.465853

Question 1.C

```
In [10]: # 4th-order Runge-Kutta
def runge_kutta(x0, y, h, x):
    # Iterate til x
    while x0 > x:
        k_1 = h * diffunc(x0,y)
        k_2 = h * diffunc(x0 + (1/2) * h, y + (1/2) * k_1)
        k_3 = h * diffunc(x0 + (1/2) * h, y + (1/2) * k_2)
        k_4 = h * diffunc(x0 + h, y + k_3)
        y = y + (1/6) * k_1 + (1/3) * k_2 + (1/3) * k_3 + (1/6) * k_4
        x0 = x0 + h
    print("Approximate solution at x = ", x, " is ", "%.6f"% y)

runge_kutta(1, -1, -0.05, 0)
runge_kutta(1, -1, -0.05, 0.05)
```

Approximate solution at x = 0 is -6071730441319061724688547840.000000Approximate solution at x = 0.05 is -5.500434

Question 1.d

```
In [73]: def inte(F,x,y,xStop,tol):
              def midpoint(F, x0, y0, xStop, nSteps):
                  h = (xStop - x0) / nSteps
                  x = x0
                  y_prev = y0
                  y_{curr} = y_{prev} + h * F(x, y_{prev})
                  for i in range(1, nSteps):
                      x = x + h
                      y_next = y_prev + 2.0 * h * F(x, y_curr)
                      y_prev = y_curr
                      y_curr = y_next
                  x = x + h
                  y_{end} = 0.5 * (y_{curr} + y_{prev} + h * F(x, y_{curr}))
                  return y end
              def richardson(r, nSteps_list):
                  k = len(r) - 1
                  for j in range(k, 0, -1):
                      factor = (nSteps_list[k] / nSteps_list[j - 1]) ** 2 - 1
                      r[j - 1] = r[j] + (r[j] - r[j - 1]) / factor
                  return
              # Maximum iteration
              kMax = 8
              \#nSteps\_list = [2 ** k for k in range(1, kMax + 1)]
              # Substep size
             nSteps list = [2*(k+1) \text{ for } k \text{ in } range(0, kMax)]
              r = []
              for nSteps in nSteps_list:
                  y_mid = midpoint(F, x, y, xStop, nSteps)
                  r.append(y_mid)
                  if len(r) > 1:
                      richardson(r, nSteps_list[:len(r)])
                      error = abs(r[0] - r[1])
                      if error < tol:</pre>
                          return r[0]
              print("Warning: Maximum iterations reached without convergence.")
              return r[0]
         def bulStoer(F, x, y, xStop, H, tol=1.0e-6):
             X = [x]
             Y = [y]
              while abs(x - xStop) > 1e-10:
                  H_{step} = min(H, xStop - x) if x < xStop else max(H, xStop - x)
                  y = inte(F, x, y, x + H_step, tol)
                  x = x + H_step
                  X.append(x)
                  Y.append(y)
              return array(X), array(Y)
         def F(x, y):
              return 1 / (x ** 2 * (1 - y))
          x0 = 1
         y0 = -1
         xStop = 0.05
         H = -0.05
          tol = 1e-6
```

```
X, Y = bulStoer(F, x0, y0, xStop, H, tol)
print("Approximate solution at x =", xStop, "is", Y[-1])
```

Approximate solution at x = 0.05 is -5.48074058058637

Question 2

```
In [389...
          def plot 3D(data):
              fig = plt.figure(figsize=(10, 8))
              ax = fig.add_subplot(111, projection='3d')
              ax.scatter(data[:, 0], data[:, 1], data[:, 2], c='blue', marker='o', s=10)
              ax.set xlabel('X')
              ax.set_ylabel('Y')
              ax.set_zlabel('Z')
              ax.set_title('3D Point Cloud Data')
              plt.show()
          def plot 2D(data, title='Projection', label="Projected Points"):
              plt.figure(figsize=(8, 6))
              plt.scatter(data[:, 0], data[:, 1], color='blue', alpha=0.7, label=label)
              plt.xlabel("X")
              plt.ylabel("Y")
              plt.title(title)
              plt.legend()
              plt.grid(True)
              plt.show()
```

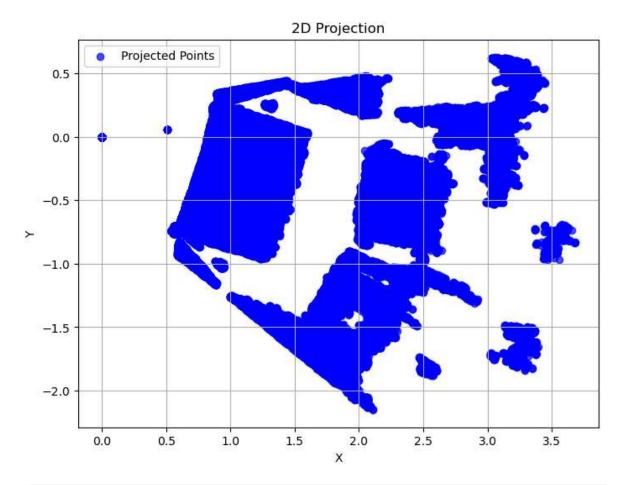
Empty Table

```
In [391...
          def PCA(data):
              # Normalize
              pc = data - np.mean(data, axis=0)
              # Covariance Matrix
              Q = np.dot(pc.T, pc) / (pc.shape[0] - 1)
              U, Sigma, Vt = np.linalg.svd(Q)
              # Eigenvectors & Eigenvalue
              eigenvectors = U
              eigenvalues = Sigma
              # Normal Vector of plane
              normal_vector = eigenvectors[:, np.argmin(eigenvalues)]
              # Principle Componenet
              largest vector = eigenvectors[:, np.argmax(eigenvalues)]
              second_largest_vector = eigenvectors[:, np.argsort(eigenvalues)[-2]]
              principal_component = np.hstack((largest_vector[:, np.newaxis], second_large
              return normal_vector, principal_component
          def PlaneParam(normal vector,data):
              Given a normal vector and data of a plane, return
              the parameter of the plane
              centorid = np.mean(data, axis=0)
              a, b, c = normal_vector
              d = - (a * centorid[0] + b * centorid[1] + c * centorid[2])
              return a, b, c, d
```

```
def PlaneProject(principle_component, data):
    Given a principle component of data, make projection
    of data into the plane of principle component
    return np.dot(data, principal components)
def planeFit(data):
    normal, principle = PCA(data)
    a,b,c,d = PlaneParam(normal, data)
    pc project = PlaneProject(principle,data)
    print('Major Plane Parameter')
    print(f'a:{a}')
    print(f'b:{b}')
    print(f'c:{c}')
    print(f'd:{d}')
    plot_2D(pc_project)
def voxel_downsample(data, voxel_size=0.02):
    coords = np.floor(data / voxel_size).astype(int)
    unique_coords, indices = np.unique(coords, axis=0, return_index=True)
    return data[indices]
def statistical_outlier_removal(data, nb_neighbors=20, std_ratio=2.0):
    tree = cKDTree(data)
    distances, _ = tree.query(data, k=nb_neighbors)
    mean_distances = np.mean(distances, axis=1)
    threshold = np.mean(mean_distances) + std_ratio * np.std(mean_distances)
    mask = mean_distances < threshold</pre>
    return data[mask]
```

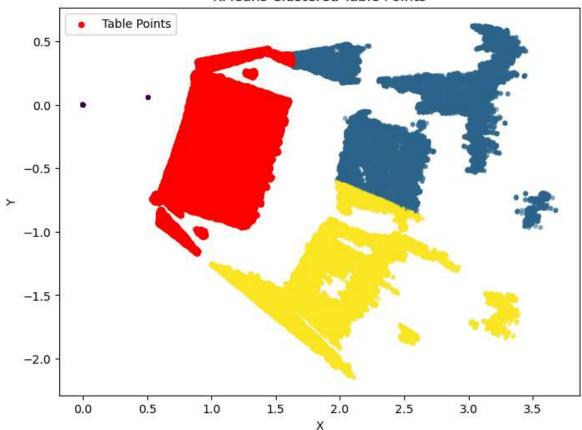
Below are my first attempt, just perform PCA and plane fitting on dominant plane

```
pc raw = np.loadtxt('./Empty2-1.asc')
In [344...
          normal, principle = PCA(pc raw)
          a,b,c,d = PlaneParam(normal, pc raw)
          pc_project = PlaneProject(principle,pc_raw)
          print('Major Plane Parameter')
          print(f'a:{a}')
          print(f'b:{b}')
          print(f'c:{c}')
          print(f'd:{d}')
          plot_2D(pc_project)
         Major Plane Parameter
         a:0.08751468415007803
         b:0.9631746638292203
         c:0.2542356131929213
         d:-0.3031026814273146
```



```
In [346...
          from sklearn.cluster import KMeans
          n_{clusters} = 4
          kmeans = KMeans(n_clusters=n_clusters, random_state=0)
          # I tried to use KMEAN to find table
          labels = kmeans.fit_predict(pc_project)
          label_counts = Counter(labels)
          table_label = label_counts.most_common(1)[0][0]
          table_points = pc_project[labels == table_label]
          plt.figure(figsize=(8, 6))
          plt.scatter(pc_project[:, 0], pc_project[:, 1], c=labels, cmap='viridis', marker
          plt.scatter(table_points[:, 0], table_points[:, 1], color='red', marker='o', s=2
          plt.xlabel("X")
          plt.ylabel("Y")
          plt.title("KMeans Clustered Table Points")
          plt.legend()
          plt.show()
```

KMeans Clustered Table Points



Below is a modified method

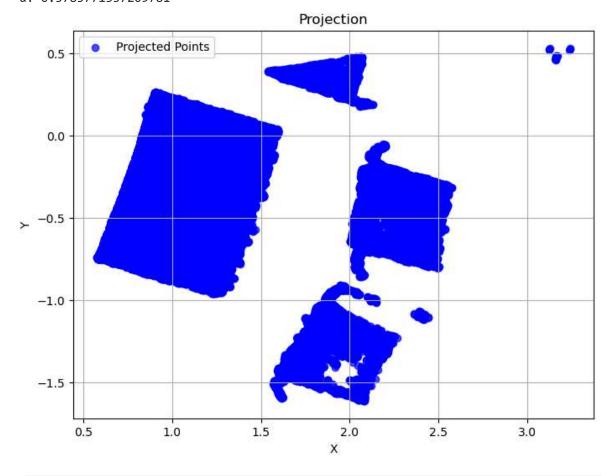
- Use RANSAC to find fit multiple plane from the point cloud, and only extract the inlier of plane with the most points
- Use DBSCAN clustering to find the table from the inlier
- Do PCA and make projection of table

```
In [404...
          import numpy as np
          from scipy.spatial import cKDTree
          import matplotlib.pyplot as plt
          from mpl toolkits.mplot3d import Axes3D
          from sklearn.linear model import RANSACRegressor
          from sklearn.preprocessing import StandardScaler
          from sklearn.cluster import DBSCAN
          def findLargestPlane(data, min_points=50, residual=0.1):
              inliers list = []
              min_points_for_plane = min_points
              while len(data) > min_points_for_plane:
                  scaler = StandardScaler()
                  data_scaled = scaler.fit_transform(data)
                  # Plan fitting
                  X = data_scaled[:, :2]
                  y = data_scaled[:, 2]
                  ransac = RANSACRegressor(residual_threshold=residual, max_trials=1000)
                  ransac.fit(X, y)
                  # Extract inlier for each plan
```

```
inlier_mask = ransac.inlier_mask_
        if np.sum(inlier_mask) < min_points_for_plane:</pre>
            break
        inliers = data[inlier mask]
        inliers list.append(inliers)
        data = data[~inlier mask]
    return inliers_list[0]
def findLargestCluster(inlier, eps=0.05, min_point=10):
    inlier scaled = StandardScaler().fit transform(inlier)
    # Apply DBSCAN clustering
    dbscan = DBSCAN(eps=0.05, min samples=10)
    labels = dbscan.fit_predict(inlier_scaled)
    # Count the number of points in each cluster
    unique_labels, counts = np.unique(labels, return_counts=True)
    largest_cluster_label = unique_labels[np.argmax(counts)]
    largest_cluster_points = inlier[labels == largest_cluster_label]
    return largest_cluster_points
```

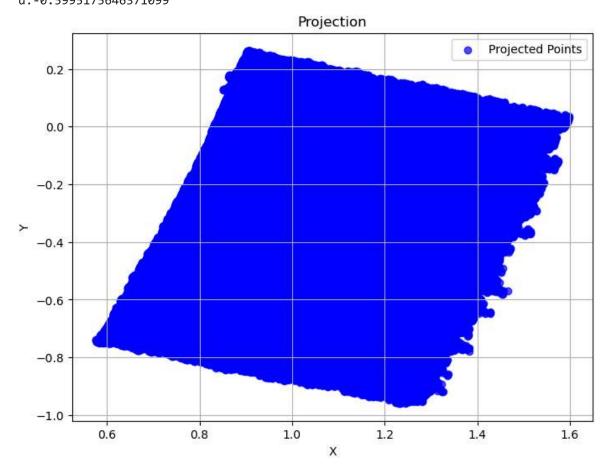
```
In [393... data = np.loadtxt('./Empty2-1.asc')
   plane = findLargestPlane(data, 50, 0.1)
   planeFit(plane)
```

Major Plane Parameter a:0.002672332282294145 b:0.9115026590987191 c:0.41128549828086247 d:-0.5785771337209781



```
In [399... table = findLargestCluster(plane, 0.05, 10)
planeFit(table)
```

Major Plane Parameter a:0.0007352700619255972 b:0.9034278478956234 c:0.4287395281807107 d:-0.5995175646371099



Clustered Tabe

I Switched to MATLAB for the rest questions, I put my MATLAB codes in the cell to make PDF

```
In [ ]: <mark>%</mark> Read Data
        data = readmatrix('TableWithObjects2-1.asc');
        ptCloud = pcdenoise(pointCloud(data));
        pointData = ptCloud.Location;
        % Parameters
        % Initial Distance & Angle Threshold
        initialDistanceThreshold = 0.2;
        initialAngleThreshold = 50;
        % Minimum Distance & Angle Threshold
        minDistanceThreshold = 0.052;
        minAngleThreshold = 5;
        % Max times of iteration
        maxIterations = 10;
        tolerance = 1e-3;
         currentDistanceThreshold = initialDistanceThreshold;
        currentAngleThreshold = initialAngleThreshold;
```

```
for i = 1:maxIterations
    % PCA: Compute Dominate Plane
    [coeff, ~, ~, ~, mu] = pca(pointData);
    normal vector = coeff(:, 3);
    % Distance of each point to Dominate Plane
    point cloud centered = pointData - mu;
    distances = abs(point_cloud_centered * normal_vector);
    % Angle Difference of each point to Dominate Plane
    normals = pcnormals(pointCloud(pointData), 10);
    angleDifferences = acosd(dot(normals, repmat(normal vector', size(normals, 1
    % Filterout outliers
    inliers = pointData((distances < currentDistanceThreshold) & (angleDifference</pre>
    % Check Convergence
    if size(inliers, 1) / size(pointData, 1) > (1 - tolerance)
        break;
    end
    pointData = inliers;
    % Reduce Angle & Distance Threshold
    currentDistanceThreshold = max(currentDistanceThreshold * 0.8, minDistanceTh
    currentAngleThreshold = max(currentAngleThreshold * 0.8, minAngleThreshold);
end
% Cluster remining inliers
epsilon = 0.3;
minPts = 10;
labels = dbscan(inliers, epsilon, minPts);
uniqueLabels = unique(labels);
% Table should be the clusters with the most dense points
maxClusterLabel = mode(labels(labels >= 0));
inliers = inliers(labels == maxClusterLabel, :);
% Plot point cloud
figure;
scatter3(inliers(:, 1), inliers(:, 2), inliers(:, 3), 10, inliers(:, 3), 'filled
xlabel('X');
ylabel('Y');
zlabel('Z');
title(['Filtered Point Cloud with ' num2str(maxIterations) ' iterations']);
axis equal;
grid on;
% Compute Table inlier
[finalCoeff, ~, ~, ~, ~, finalMu] = pca(inliers);
planeNormal = finalCoeff(:, 3);
planePoint = finalMu;
a = planeNormal(1);
b = planeNormal(2);
c = planeNormal(3);
d = -dot(planeNormal, planePoint);
fprintf('Table Parameter: %fx + %fy + %fz + %f = 0 \ n', a, b, c, d);
```

CSE

```
In [ ]: data = readmatrix('CSE-1.asc');
        ptCloud = pcdenoise(pointCloud(data));
        pointData = ptCloud.Location;
        % Find Floor
        % PCA: Compute Floor Plane
         [coeff, ~, ~, ~, ~, mu] = pca(pointData);
        normal vector = coeff(:, 3);
        maxDistance = 0.5;
        referenceVector = normal vector;
        maxAngularDistance = 5;
        % Plane Fitting: Floor Plane
         [model1,inlierIndices,outlierIndices] = pcfitplane(ptCloud,...
            maxDistance, referenceVector, maxAngularDistance);
        plane1 = select(ptCloud,inlierIndices);
        remainPtCloud = select(ptCloud,outlierIndices);
        % Find First Wall
         % PCA: Compute First Wall
         remaining_points = remainPtCloud.Location;
         [wall_coeff, ~, ~, ~, ~, ~] = pca(remaining_points);
        % First Wall Normal: Perpendicular to floor
        angles = zeros(3,1);
        floor_normal = normal_vector;
        for i = 1:3
            angles(i) = abs(dot(wall_coeff(:,i), floor_normal));
        end
         [~, idx] = min(angles);
        wall_reference = wall_coeff(:, idx);
        wall_reference = wall_reference - (dot(wall_reference, floor_normal) * floor_nor
        wall_reference = wall_reference / norm(wall_reference);
        % Plane Fitting: First Wall
        maxDistance = 0.6;
        maxAngularDistance = 10;
         [wallModel, wallInliers, wallOutliers] = pcfitplane(remainPtCloud, ...
            maxDistance, wall reference, maxAngularDistance);
        wall1 = select(remainPtCloud, wallInliers);
        % Find Remaining Wall and Objects
         remaining_points = select(remainPtCloud, wallOutliers);
        wall planes = {};
        minPoints = 500;
        wall plane param = {};
         for iter = 1:5
            maxDistance = 0.5;
            maxAngularDistance = 10;
            % Wall & Object Normal: Perpendicular to First Wall and Floor
            referenceVector = cross(floor_normal, first_wall_normal);
            referenceVector = referenceVector / norm(referenceVector);
            % Plane Fitting
            [model, inlierIndices, outlierIndices] = pcfitplane(remaining_points, ...
                 maxDistance, referenceVector, maxAngularDistance);
            if numel(inlierIndices) < minPoints</pre>
                 break;
            new_wall = select(remaining_points, inlierIndices);
```

```
wall_planes{end+1} = new_wall;
    wall_plane_param{end+1} = model;
    % Remove found Wall & Object From the cloud
    validOutliers = outlierIndices(outlierIndices <= remaining points.Count);</pre>
    remaining points = select(remaining points, validOutliers);
end
% Visualize
figure;
% Floor
pcshow(plane1.Location, 'g');
hold on;
% First Wall
pcshow(wall1.Location, 'r');
fprintf('Parameters of red wall:\n');
disp(wallModel);
% Each other Objects & Wall will be assigned with random color
for i = 1:length(wall_planes)
    pcshow(wall_planes{i}.Location, rand(1, 3));
    wall_center = mean(wall_planes{i}.Location, 1);
    text(wall_center(1), wall_center(2), wall_center(3), sprintf('Wall %d', i),
         'Color', 'r', 'FontSize', 12, 'FontWeight', 'bold');
    fprintf('Parameters of wall %d:\n', i);
    disp(wall_plane_param{i});
end
% Remaining Point Will be Black
pcshow(remaining_points.Location, 'k');
title('Ground and Walls');
xlabel('X');
ylabel('Y');
zlabel('Z');
legend('Floor', 'Wall 1', 'Additional Walls', 'Remaining Points');
```